

Crop recommendation model

```
# Importing libraries
```

```
from __future__ import print_function
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import classification_report
from sklearn import metrics
from sklearn import tree
import warnings
warnings.filterwarnings('ignore')
```

In [12]:

```
PATH = '/content/Crop_recommendation.csv'
df = pd.read_csv(PATH)
```

In [13]:

```
df.head()
```

Out[13]:

	N	P	K	temperature	humidity	ph	rainfall	label
0	90	42	43	20.879744	82.002744	6.502985	202.935536	rice
1	85	58	41	21.770462	80.319644	7.038096	226.655537	rice
2	60	55	44	23.004459	82.320763	7.840207	263.964248	rice
3	74	35	40	26.491096	80.158363	6.980401	242.864034	rice
4	78	42	42	20.130175	81.604873	7.628473	262.717340	rice

In [14]:

```
df.tail()
```

Out[14]:

	N	P	K	temperature	humidity	ph	rainfall	label
2195	107	34	32	26.774637	66.413269	6.780064	177.774507	coffee
2196	99	15	27	27.417112	56.636362	6.086922	127.924610	coffee
2197	118	33	30	24.131797	67.225123	6.362608	173.322839	coffee
2198	117	32	34	26.272418	52.127394	6.758793	127.175293	coffee

	N	P	K	temperature	humidity	ph	rainfall	label
2199	104	18	30	23.603016	60.396475	6.779833	140.937041	coffee

```
df.size
```

In [15]:

```
17600
```

Out[15]:

```
df.shape
```

In [16]:

```
(2200, 8)
```

Out[16]:

```
df.columns
```

In [17]:

```
Index(['N', 'P', 'K', 'temperature', 'humidity', 'ph', 'rainfall', 'label'], dtype='object')
```

Out[17]:

```
df['label'].unique()
```

In [18]:

```
array(['rice', 'maize', 'chickpea', 'kidneybeans', 'pigeonpeas',
       'mothbeans', 'mungbean', 'blackgram', 'lentil', 'pomegranate',
       'banana', 'mango', 'grapes', 'watermelon', 'muskmelon', 'apple',
       'orange', 'papaya', 'coconut', 'cotton', 'jute', 'coffee'],
      dtype=object)
```

Out[18]:

```
df.dtypes
```

In [19]:

```
N          int64
P          int64
K          int64
temperature float64
humidity    float64
ph          float64
rainfall    float64
label       object
dtype: object
```

Out[19]:

```
df['label'].value_counts()
```

In [20]:

```
rice          100
maize         100
jute          100
cotton        100
coconut       100
papaya        100
orange        100
apple         100
muskmelon     100
watermelon    100
grapes        100
```

Out[20]:

```

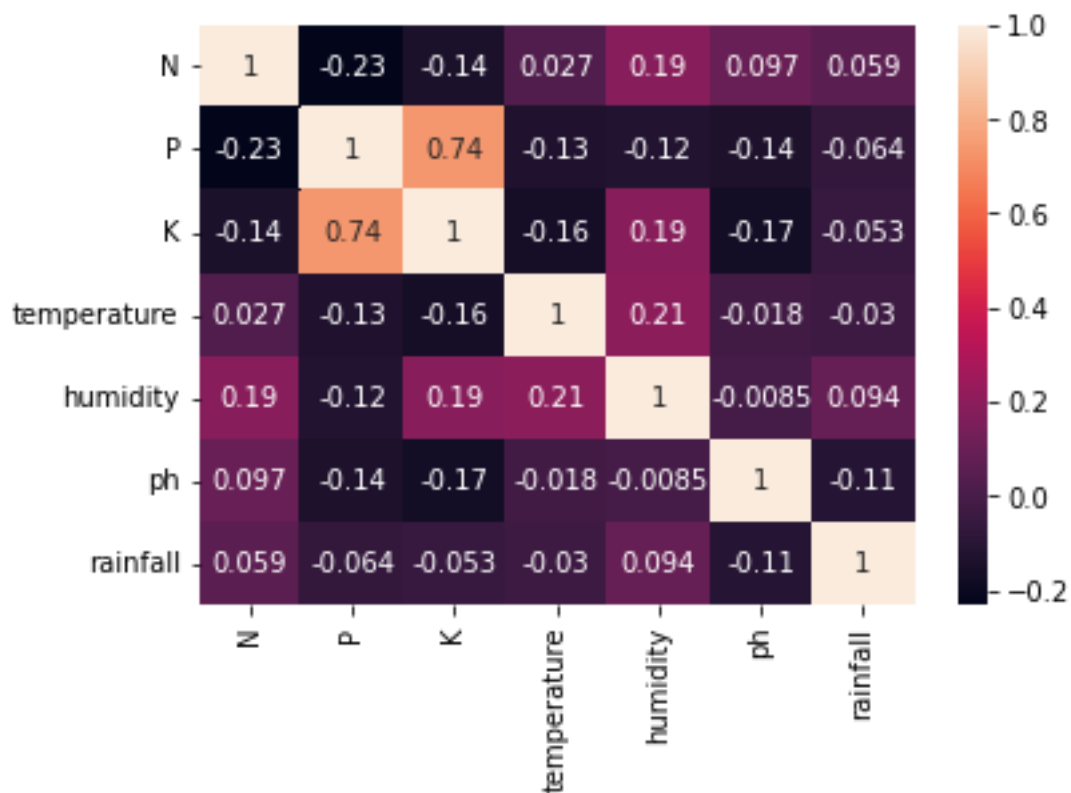
mango          100
banana         100
pomegranate    100
lentil         100
blackgram      100
mungbean       100
mothbeans      100
pigeonpeas     100
kidneybeans    100
chickpea       100
coffee        100
Name: label, dtype: int64

```

In [21]:

```
sns.heatmap(df.corr(),annot=True)
```

Out[21]:



Seperating features and target label

In [22]:

```

features = df[['N', 'P','K','temperature', 'humidity', 'ph', 'rainfall']]
target = df['label']
labels = df['label']

```

In [23]:

```

# Initializing empty lists to append all model's name and corresponding
name
acc = []
model = []

```

In [24]:

```

from sklearn.model_selection import train_test_split
Xtrain, Xtest, Ytrain, Ytest = train_test_split(features,target,test_size =
0.2,random_state =2)

```

Decision Tree

In [25]:

```
from sklearn.tree import DecisionTreeClassifier

DecisionTree =
DecisionTreeClassifier(criterion="entropy", random_state=2, max_depth=5)

DecisionTree.fit(Xtrain, Ytrain)

predicted_values = DecisionTree.predict(Xtest)
x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Decision Tree')
print("DecisionTrees's Accuracy is: ", x*100)

print(classification_report(Ytest, predicted_values))
DecisionTrees's Accuracy is:  90.0
```

	precision	recall	f1-score	support
apple	1.00	1.00	1.00	13
banana	1.00	1.00	1.00	17
blackgram	0.59	1.00	0.74	16
chickpea	1.00	1.00	1.00	21
coconut	0.91	1.00	0.95	21
coffee	1.00	1.00	1.00	22
cotton	1.00	1.00	1.00	20
grapes	1.00	1.00	1.00	18
jute	0.74	0.93	0.83	28
kidneybeans	0.00	0.00	0.00	14
lentil	0.68	1.00	0.81	23
maize	1.00	1.00	1.00	21
mango	1.00	1.00	1.00	26
mothbeans	0.00	0.00	0.00	19
mungbean	1.00	1.00	1.00	24
muskmelon	1.00	1.00	1.00	23
orange	1.00	1.00	1.00	29
papaya	1.00	0.84	0.91	19
pigeonpeas	0.62	1.00	0.77	18
pomegranate	1.00	1.00	1.00	17
rice	1.00	0.62	0.77	16
watermelon	1.00	1.00	1.00	15
accuracy			0.90	440
macro avg	0.84	0.88	0.85	440
weighted avg	0.86	0.90	0.87	440

In [26]:

```
from sklearn.model_selection import cross_val_score
```

In [27]:

```
# Cross validation score (Decision Tree)
score = cross_val_score(DecisionTree, features, target, cv=5)
```

In [28]:

```
score
```

Out[28]:

```
array([0.93636364, 0.90909091, 0.91818182, 0.87045455, 0.93636364])
```

Saving trained Decision Tree model

In [29]:

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
DT_pkl_filename = 'DecisionTree.pkl'
# Open the file to save as pkl file
DT_Model_pkl = open(DT_pkl_filename, 'wb')
pickle.dump(DecisionTree, DT_Model_pkl)
# Close the pickle instances
DT_Model_pkl.close()
```

In [30]:

```
from sklearn.naive_bayes import GaussianNB

NaiveBayes = GaussianNB()

NaiveBayes.fit(Xtrain,Ytrain)

predicted_values = NaiveBayes.predict(Xtest)
x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Naive Bayes')
print("Naive Bayes's Accuracy is: ", x)

print(classification_report(Ytest,predicted_values))
Naive Bayes's Accuracy is:  0.990909090909091
      precision    recall  f1-score   support

   apple          1.00      1.00      1.00        13
  banana          1.00      1.00      1.00        17
blackgram          1.00      1.00      1.00        16
 chickpea          1.00      1.00      1.00        21
  coconut          1.00      1.00      1.00        21
   coffee          1.00      1.00      1.00        22
   cotton          1.00      1.00      1.00        20
   grapes          1.00      1.00      1.00        18
     jute          0.88      1.00      0.93        28
kidneybeans          1.00      1.00      1.00        14
   lentil          1.00      1.00      1.00        23
    maize          1.00      1.00      1.00        21
    mango          1.00      1.00      1.00        26
 mothbeans          1.00      1.00      1.00        19
  mungbean          1.00      1.00      1.00        24
 muskmelon          1.00      1.00      1.00        23
   orange          1.00      1.00      1.00        29
   papaya          1.00      1.00      1.00        19
pigeonpeas          1.00      1.00      1.00        18
pomegranate          1.00      1.00      1.00        17
     rice          1.00      0.75      0.86        16
watermelon          1.00      1.00      1.00        15

 accuracy                   0.99         440
  macro avg              0.99              440
weighted avg              0.99              440
```

In [31]:

```
# Cross validation score (NaiveBayes)
score = cross_val_score(NaiveBayes, features, target, cv=5)
score
```

Out[31]:

```
array([0.99772727, 0.99545455, 0.99545455, 0.99545455, 0.99090909])
```

In [32]:

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
NB_pkl_filename = 'NBClassifier.pkl'
# Open the file to save as pkl file
NB_Model_pkl = open(NB_pkl_filename, 'wb')
pickle.dump(NaiveBayes, NB_Model_pkl)
# Close the pickle instances
NB_Model_pkl.close()
```

In [33]:

```
from sklearn.svm import SVC

SVM = SVC(gamma='auto')

SVM.fit(Xtrain, Ytrain)

predicted_values = SVM.predict(Xtest)

x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('SVM')
print("SVM's Accuracy is: ", x)

print(classification_report(Ytest, predicted_values))

SVM's Accuracy is: 0.10681818181818181
      precision    recall  f1-score   support

   apple         1.00      0.23      0.38         13
  banana         1.00      0.24      0.38         17
blackgram         1.00      0.19      0.32         16
 chickpea         1.00      0.05      0.09         21
   coconut         1.00      0.05      0.09         21
    coffee         0.00      0.00      0.00         22
    cotton         1.00      0.05      0.10         20
    grapes         1.00      0.06      0.11         18
     jute         1.00      0.07      0.13         28
kidneybeans       0.03      1.00      0.07         14
    lentil         0.00      0.00      0.00         23
     maize         0.00      0.00      0.00         21
     mango         0.00      0.00      0.00         26
  mothbeans       0.00      0.00      0.00         19
   mungbean         1.00      0.12      0.22         24
 muskmelon         1.00      0.30      0.47         23
    orange         1.00      0.03      0.07         29
   papaya         1.00      0.05      0.10         19
pigeonpeas        0.00      0.00      0.00         18
pomegranate       1.00      0.12      0.21         17
     rice         0.50      0.06      0.11         16
 watermelon       1.00      0.13      0.24         15
```

accuracy			0.11	440
macro avg	0.66	0.13	0.14	440
weighted avg	0.66	0.11	0.13	440

In [34]:

```
# Cross validation score (SVM)
score = cross_val_score(SVM, features, target, cv=5)
score
```

Out[34]:

```
array([0.27727273, 0.28863636, 0.29090909, 0.275      , 0.26818182])
```

In [35]:

```
from sklearn.linear_model import LogisticRegression

LogReg = LogisticRegression(random_state=2)

LogReg.fit(Xtrain, Ytrain)

predicted_values = LogReg.predict(Xtest)

x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('Logistic Regression')
print("Logistic Regression's Accuracy is: ", x)

print(classification_report(Ytest, predicted_values))
Logistic Regression's Accuracy is:  0.9522727272727273
      precision    recall  f1-score   support

   apple          1.00      1.00      1.00         13
  banana          1.00      1.00      1.00         17
blackgram          0.86      0.75      0.80         16
 chickpea          1.00      1.00      1.00         21
  coconut          1.00      1.00      1.00         21
   coffee          1.00      1.00      1.00         22
   cotton          0.86      0.90      0.88         20
   grapes          1.00      1.00      1.00         18
    jute          0.84      0.93      0.88         28
kidneybeans          1.00      1.00      1.00         14
   lentil          0.88      1.00      0.94         23
   maize          0.90      0.86      0.88         21
   mango          0.96      1.00      0.98         26
 mothbeans          0.84      0.84      0.84         19
 mungbean          1.00      0.96      0.98         24
 muskmelon          1.00      1.00      1.00         23
   orange          1.00      1.00      1.00         29
  papaya          1.00      0.95      0.97         19
pigeonpeas          1.00      1.00      1.00         18
pomegranate          1.00      1.00      1.00         17
    rice          0.85      0.69      0.76         16
watermelon          1.00      1.00      1.00         15

   accuracy          0.95          0.95          0.95          440
  macro avg          0.95          0.95          0.95          440
weighted avg          0.95          0.95          0.95          440
```

In [36]:

```
# Cross validation score (Logistic Regression)
score = cross_val_score(LogReg, features, target, cv=5)
score
```

Out[36]:

```
array([0.95      , 0.96590909, 0.94772727, 0.96818182, 0.94318182])
```

In [37]:

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
LR_pkl_filename = 'LogisticRegression.pkl'
# Open the file to save as pkl file
LR_Model_pkl = open(DT_pkl_filename, 'wb')
pickle.dump(LogReg, LR_Model_pkl)
# Close the pickle instances
LR_Model_pkl.close()
```

In [38]:

```
from sklearn.ensemble import RandomForestClassifier

RF = RandomForestClassifier(n_estimators=20, random_state=0)
RF.fit(Xtrain, Ytrain)

predicted_values = RF.predict(Xtest)

x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('RF')
print("RF's Accuracy is: ", x)

print(classification_report(Ytest, predicted_values))
RF's Accuracy is: 0.990909090909091
```

	precision	recall	f1-score	support
apple	1.00	1.00	1.00	13
banana	1.00	1.00	1.00	17
blackgram	0.94	1.00	0.97	16
chickpea	1.00	1.00	1.00	21
coconut	1.00	1.00	1.00	21
coffee	1.00	1.00	1.00	22
cotton	1.00	1.00	1.00	20
grapes	1.00	1.00	1.00	18
jute	0.90	1.00	0.95	28
kidneybeans	1.00	1.00	1.00	14
lentil	1.00	1.00	1.00	23
maize	1.00	1.00	1.00	21
mango	1.00	1.00	1.00	26
mothbeans	1.00	0.95	0.97	19
mungbean	1.00	1.00	1.00	24
muskmelon	1.00	1.00	1.00	23
orange	1.00	1.00	1.00	29
papaya	1.00	1.00	1.00	19
pigeonpeas	1.00	1.00	1.00	18
pomegranate	1.00	1.00	1.00	17
rice	1.00	0.81	0.90	16
watermelon	1.00	1.00	1.00	15

accuracy			0.99	440
macro avg	0.99	0.99	0.99	440
weighted avg	0.99	0.99	0.99	440

In [39]:

```
# Cross validation score (Random Forest)
score = cross_val_score(RF, features, target, cv=5)
score
```

Out[39]:

```
array([0.99772727, 0.99545455, 0.99772727, 0.99318182, 0.98863636])
```

In [40]:

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
RF_pkl_filename = 'RandomForest.pkl'
# Open the file to save as pkl file
RF_Model_pkl = open(RF_pkl_filename, 'wb')
pickle.dump(RF, RF_Model_pkl)
# Close the pickle instances
RF_Model_pkl.close()
```

In [41]:

```
import xgboost as xgb
XB = xgb.XGBClassifier()
XB.fit(Xtrain, Ytrain)

predicted_values = XB.predict(Xtest)

x = metrics.accuracy_score(Ytest, predicted_values)
acc.append(x)
model.append('XGBoost')
print("XGBoost's Accuracy is: ", x)

print(classification_report(Ytest, predicted_values))
XGBoost's Accuracy is: 0.9931818181818182
      precision    recall  f1-score   support

   apple         1.00      1.00      1.00        13
  banana         1.00      1.00      1.00        17
blackgram         1.00      1.00      1.00        16
 chickpea         1.00      1.00      1.00        21
   coconut         1.00      1.00      1.00        21
   coffee         1.00      1.00      1.00        22
   cotton         1.00      1.00      1.00        20
   grapes         1.00      1.00      1.00        18
     jute         0.96      0.93      0.95        28
kidneybeans         1.00      1.00      1.00        14
   lentil         1.00      1.00      1.00        23
   maize         1.00      1.00      1.00        21
   mango         1.00      1.00      1.00        26
 mothbeans         1.00      1.00      1.00        19
 mungbean         1.00      1.00      1.00        24
 muskmelon         1.00      1.00      1.00        23
   orange         1.00      1.00      1.00        29
   papaya         1.00      1.00      1.00        19
pigeonpeas         1.00      1.00      1.00        18
```

pomegranate	1.00	1.00	1.00	17
rice	0.88	0.94	0.91	16
watermelon	1.00	1.00	1.00	15
accuracy			0.99	440
macro avg	0.99	0.99	0.99	440
weighted avg	0.99	0.99	0.99	440

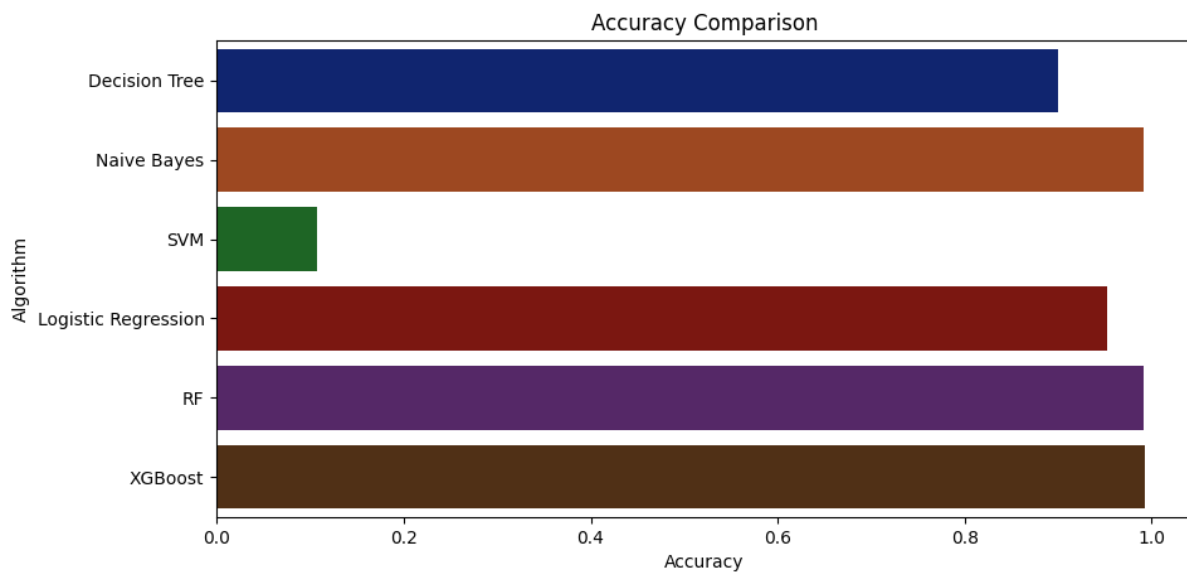
In [42]:

```
import pickle
# Dump the trained Naive Bayes classifier with Pickle
XB_pkl_filename = 'XGBoost.pkl'
# Open the file to save as pkl file
XB_Model_pkl = open(XB_pkl_filename, 'wb')
pickle.dump(XB, XB_Model_pkl)
# Close the pickle instances
XB_Model_pkl.close()
```

In [43]:

```
plt.figure(figsize=[10,5],dpi = 100)
plt.title('Accuracy Comparison')
plt.xlabel('Accuracy')
plt.ylabel('Algorithm')
sns.barplot(x = acc,y = model,palette='dark')
```

Out[43]:



In [44]:

```
accuracy_models = dict(zip(model, acc))
for k, v in accuracy_models.items():
    print (k, '-->', v)

Decision Tree --> 0.9
Naive Bayes --> 0.990909090909091
SVM --> 0.10681818181818181
Logistic Regression --> 0.9522727272727273
RF --> 0.990909090909091
XGBoost --> 0.9931818181818182
```

In [45]:

```
data = np.array([[104,18, 30, 23.603016, 60.3, 6.7, 140.91]])
prediction = RF.predict(data)
```

```
print(prediction)
```

```
['coffee']
```

```
data = np.array([[83, 45, 60, 28, 70.3, 7.0, 150.9]])
```

```
prediction = RF.predict(data)
```

```
print(prediction)
```

```
['jute']
```

In [46]: